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A PROBABILISTIC ASSESSMENT OF NASA ULTRA-EFFICIENT ENGINE TECHNOLOGIES FOR A LARGE SUBSONIC TRANSPORT

Michael T. Tong and Scott M. Jones

National Aeronautics and Space Administration Glenn Research Center Cleveland, Ohio 44135

Philip C. Arcara, Jr.

National Aeronautics and Space Administration Langley Research Center Hampton, Virginia 23681

William J. Haller

National Aeronautics and Space Administration Glenn Research Center Cleveland, Ohio 44135

ABSTRACT

NASA's Ultra Efficient Engine Technology (UEET) program features advanced aeropropulsion technologies that include highly loaded turbomachinery, an advanced low-NO_x combustor, high-temperature materials, intelligent propulsion controls, aspirated seal technology, and an advanced computational fluid dynamics (CFD) design tool to help reduce airplane drag. A probabilistic system assessment is performed to evaluate the impact of these technologies on aircraft fuel burn and NO_x reductions. A 300-passenger aircraft, with two 396-kN thrust (85,000-pound) engines is chosen for the study. The results show that a large subsonic aircraft equipped with the UEET technologies has a very high probability of meeting the UEET Program goals for fuel-burn (or equivalent CO₂) reduction (-15% from the baseline) and LTO (landing and takeoff) NO_x reductions (-70% relative to the 1996 International Civil Aviation Organization rule). These results are used to provide guidance for developing a robust UEET technology portfolio, and to prioritize the most promising technologies required to achieve UEET program goals for the fuel-burn and NO_x reductions.

Keywords: NASA, ultra-efficient engine technologies, probabilistic, propulsion system analysis, fuel burn, CO_2 , LTO NO_x

INTRODUCTION

Throughout the past century, propulsion innovations were the driving force behind the evolution of air transportation. Advances in propulsion system technology offer the greatest single contribution to the improvement of fuel economy, capacity, and the environmental impact of commercial aircraft.

In the twenty-first century, propulsion will continue to be the enabling technology to revolutionize air transportation. To maintain the superiority of U.S. aerospace industry into the 21st century, and to assist in meeting public demand for safety, reliability, environmental compatibility, and affordability, NASA initiated the Ultra-Efficient Engine Technology (UEET) program in year 2000.

NASA's UEET program is designed to revolutionize the state of the art in propulsion and propulsion/airframe integration technologies with specific objectives to reduce aircraft fuel burn (or CO_2) and NO_x , relative to current engines. It features advanced technologies that include:

Tech ID	Technology name
tech-1	advanced low NO _x combustor
tech-2	highly loaded compressor technology
tech-3	highly loaded turbine systems
tech-4	1482 °C (2700 °F) ceramic matrix composite
	(CMC) turbine vane
tech-5	1482 °C (2700 °F) CMC combustor liner
tech-6	advanced compressor disk alloy
tech-7	low conductivity ceramic thermal barrier
	coating (TBC) for turbine airfoils
tech-8	advanced airfoil alloy development
tech-9	high Reynolds number (advanced CFD)
	design tool for advanced configuration
tech-10	rotating machinery clearance management
tech-11	high temperature wireless data communication
tech-12	aspirating seal

The NASA UEET Program goals are a 70% reduction in landing and take-off (LTO) NO_x relative to the 1996 International Civil Aviation Organization (ICAO) standard and a 15% fuel-burn improvement relative to the current state of the art large subsonic transports.

A probabilistic system assessment is performed to evaluate the impact of these technologies on aircraft fuel-burn and NO_x reductions. The statistical approach quantifies the uncertainties inherent in these new propulsion technologies and their influence on the likely outcomes of engine performance. Consequently, it provides additional insight into the risks associated with new technologies, which are often needed by the decision-makers to determine the benefit and return-on-investment of new propulsion technologies. The results are used to provide guidance for the development of a robust UEET technology portfolio, and to prioritize the most promising technologies required to achieve UEET program goals for the fuel-burn and NO_x reductions.

ANALYSIS APPROACH AND PROCEDURES

Expert Opinion Elicitation

As Rand analyst E.S. Quade observed about 30 years ago, "Intuition and judgment permeate all analysis ... As questions get broader, intuition and judgment must supplement quantitative analysis to an increasing extent" [1]. Expert judgment must be used to judge the risks of emerging technology. Obtaining and quantifying input data is probably the most crucial part of performing risk assessments. It is a crucial but generally overlooked issue [2]. As such, it deserves detailed attention.

For the current assessment, a technology audit scheme (TAS) developed by Kirby and Mavris [3] is used to elicit opinions from the NASA technologists identified as the focal point for each of the UEET technologies. The scheme is based on the Delphi method [4], which is a structured process for collecting and distilling knowledge from a group of experts by means of a series of questionnaires and interviews interspersed with controlled opinion feedback. The focus of the TAS is to identify the applicable set of UEET technologies for the vehicle of interest, gather the required information, and compile the data necessary for the system analysis. The process is described in detail in reference [5].

The Beta Distribution

Based on the information obtained from the technologists, the 3-point estimates (maximum, minimum, and most-likely values) of the impacts (positive and/or negative) for each of the technologies are quantified. They are summarized in Table 1. A four-parameter beta distribution is then created for each of the technologies. The probability density function (PDF) of the beta distribution is

$$f(x) = \frac{1}{B(p,q)} \frac{(x-a)^{p-1} (b-x)^{q-1}}{(b-a)^{p+q-1}} \tag{1}$$

and the cumulative density function (CDF) is

$$CDF(t) = \frac{1}{B(p,q)} \int_{0}^{t} t^{p-1} (1-t)^{q-1} dt$$
 (2)

with the transformation

$$t = \frac{(x-a)}{(b-a)}$$

where the parameters a and b are the minimum and maximum values of the variable x, respectively; p and q are the distribution shape parameters and B is the beta function defined by

$$B(p,q) = \frac{\Gamma(p) \cdot \Gamma(q)}{\Gamma(p+q)} = \int_{0}^{1} t^{p-1} (1-t)^{q-1} dt$$
 (3)

The shape parameters p and q depend on whether the mode (most-likely value) is to the left or right of the midrange. They are determined using the method described in [6]. The resulted mean and standard deviation of the impact for each of the technologies are also summarized in Table 1.

The CDFs (eq. (2)) are calculated numerically. All three equations are implemented into the Fast Probability Integration (FPI) computer code [7], and are used to perform the probabilistic system analysis of the UEET technologies.

System Analysis

The approach taken in this effort is to combine thermodynamic cycle analysis using NPSS (Numerical Propulsion System Simulator) [8], engine weight estimation using WATE (Weight Analysis of Gas Turbine Engines) [9], aircraft mission sizing using FLOPS (Flight Optimization System) [10], and FPI. A schematic of the integrated approached is shown in Figure 1.

The computer code NPSS is used to calculate engine thrust, specific fuel consumption and LTO NO_x emissions. The engine weight is calculated by the WATE code. The results from NPSS and WATE are used by FLOPS for performing airplane mission and sizing analyses, and ultimately calculates the fuel-burn based on a 5556-kilometers (3000 nautical miles) economic mission.

Probabilistic Analysis

All probabilistic analysis methods are approximate. Monte Carlo simulation, which is oftentimes referred to as the "exact" solution, is actually an approximate because a finite number of samples are always used. Thus, the nature of the approximation is one of "lack of data", which can be reduced by increasing the number of samples. However, for large-scale high fidelity problems, the inefficiency of Monte Carlo simulation renders it impractical for use. Many efficient methods have been developed to alleviate the need for Monte Carlo simulation. These methods include the first and second-order reliability method (FORM and SORM) [11], the advanced mean value family of methods (AMV) (12], and the response surface method (RSM) [13]. These methods replace the original deterministic model with a computationally efficient analytical model in order to speed up the analysis.

1:

Tal	ble 1.—UEET tech	inologies and their ur	Table 1.—UEET technologies and their uncertainties for a large subsonic transport (based on year-2003 technology audit)	subsonic transport	(based on year-2003 t	echnology audii	.;
Technology identification	Technology	Maximum impact	Minimum impact	Most-likely impact	Mean impact	Standard deviation	Distribution
tech-1	Advanced low NO _x combustor	75% LTO NO _x reduction	65% LTO NOx reduction	70% LTO NOx reduction	70% LTO NOx reduction	1.67%	Beta
tech-2	Highly-loaded compressor	+45% HPC work factor +1 pt HPC poly. eff.	+27% HPC work factor	+38% HPC work factor	+37.4% HPC work factor +0.2 pt. poly. eff.	3.16% 0.16 pt.	Beta
tech-3a*	Highly-loaded HP turbine	+21% HPT loading +0.5 pt. poly. eff.	+19% HPT loading -0.5 pts. poly. eff.	+20% HPT loading	+20% HPT loading +0.0 pt. poly. eff.	0.33% 0.17 pt.	Beta
tech-3b*	Highly-loaded LP turbine	+30% LPT loading +3 pts. LPT poly. eff. +0.5% HPC bleed	+25% LPT loading +1.0 pt. poly. eff. +2.0% HPC bleed	+28% LPT loading +2 pts. poly. eff. +0.5% HPC bleed	+27.9% LPT loading +2 pts. poly. eff. +0.8% HPC bleed	0.87% 0.33 pt. 0.24%	Beta
tech-4	1482 °C (2700 °F) CMC turbine vane	+389 °C (700 °F) HPT vane temp. CMC 1st stage HPT vane	+361 °C (650 °F) HPT vane temp.	+389 °C (700 °F) HPT vane temp. CMC 1st stage HPT vane	+383 °C (690 °F) HPT vane temp. CMC 1st stage HPT vane	4.6 °C (8.2 °F)	Beta
tech-5	1482 °C (2700 °F) CMC combustor liner	reduce cooling flow by 60%	reduce cooling flow by 53%	reduce cooling flow by 57%	reduce cooling flow by 57%	1.21%	Beta
tech-6	Advanced compressor alloy	OPR up to 1005 K (1810 R) T ₃	OPR up to 995 K (1760 R) T ₃	OPR up to 1005 K (1810 R) T ₃	OPR up to 1000 K (1800 R) T ₃	4.6 K (8.2 R)	Beta
tech-7	Low conductivity thermal barrier coating for turbine airfoil	+139 K (250 R) HPT and LPT blade and vane temp. (reduce cooling flow)	+83 K (150 R) HPT and LPT blade and vane temp. (reduce cooling flow)	+111 K (200 R) HPT and LPT blade and vane temp. (reduce cooling flow)	+111 K (200 R) HPT and LPT blade and vane temp. (reduce cooling flow)	9.3 K (16.7 R)	Beta
tech-8a*	Advanced airfoil alloy development for HPT	+56 K (100 R) HPT blade and vane temp. (reduce cooling flow); —3.85% HPT blade and vane densities	+23 K (42 R) HPT blade and vane temp. (reduce cooling flow)	+43 K (78 R) HPT blade and vane temp. (reduce cooling flow) –2.24% HPT blade and vane densities	+42 K (76 R) HPT blade and vane temp. (reduce cooling flow) -2.15% HPT blade and vane densities	5.7 K (10.2 R) 0.67%	Beta
tech-8b*	Advanced airfoil alloy development for LPT	+57 K (102 R) LPT blade and vane temp. (reduce cooling flow); —4.15% LPT blade and vane densities	+40 K (72 R) LPT blade and vane temp. (reduce cooling flow) -0.32% LPT blade and vane densities	+52 K (94 R) LPT blade and vane temp. (reduce cooling flow); –2.56% LPT blade and vane densities	+51 K (92 R) LPT blade and vane temp. (reduce cooling flow); –2.47% LPT blade and vane densities	2.9 K (5.3 R) 0.67%	Beta
tech-9	High Reynolds number design tool	-1.30% aircraft drag	-0.20% aircraft drag	-1.30% aircraft drag	-1.08% aircraft drag	0.18%	Beta
tech-10	Rotational machine clearance management	+1.00 pt HPT adia. eff. +0.20 pt. LPT adia. eff. +0.50 pt HPC poly. eff.	+0.50 pt HPT adia. eff. +0.25 pt HPC poly. eff. +34 kg (75 lbs) engine wt.	+0.75 pt HPT adia. eff. +0.10 pt. LPT adia. eff. +0.40 pt HPC poly. eff.	+0.75 pt HPT adia. eff. +0.1 pt. LPT adia. eff. +0.39 pt HPC poly. eff. +7 kg (15 lbs) engine wt.	0.08 pt. 0.03 pt. 0.04 pt. 5.6 kg (12.3 lbs)	Beta
tech-11	High-temperature wireless data communication	-113 kg (-250 lbs) engine wt.	–48 kg (–105 lbs) engine wt.	-77 kg (-170 lbs) engine wt.	–78 kg (–172 lbs) engine wt.	11 kg (25 lbs)	Beta
tech-12	Aspiring seal	+0.775 pt HPT adia. eff. +5 kg (10 lbs) HPT wt.	12 Aspiring seal +0.775 pt HPT adia. eff. +0.625 pt HPT adia. eff. +0.750 pt HPT adia. eff. +0.732 pt HPT adia. eff. 0.03 pt. Beta +7 kg (15 lbs) HPT wt. +7 kg (15 lbs) HPT wt. +7 kg (15 lbs) HPT wt. 0.8 kg (1.7 lbs) Beta	+0.750 pt HPT adia. eff. +7 kg (15 lbs) HPT wt.	+0.732 pt HPT adia. eff. +7 kg (15 lbs) HPT wt.	0.03 pt. 0.8 kg (1.7 lbs)	Beta
*Note: re	esults of tech-3a and tec	ch-3b are combined to show	the benefit of highly-loaded tu	irbine system technology (†	ech-3): results of tech-8a and	18b are combined to	show the benef

*Note: results of tech-3a and tech-3b are combined to show the benefit of highly-loaded turbin of advanced airfoil alloy development technology (tech-8).

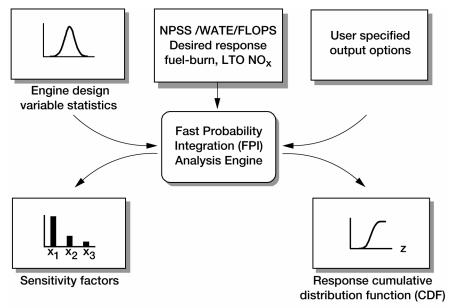


Figure 1.—Fast probability integration input/output schematic.

For the current assessment, an advanced first-order reliability method is used. This method, based on the most-probable-point (MPP) concept, is one of the several methods in the FPI code. The code was developed under contract with NASA Glenn Research Center [14]. The role of FPI is to perform probabilistic analysis utilizing the results generated by NPSS, WATE, and FLOPS. The results are generated in the form of cumulative distribution functions (CDFs).

In addition, FPI is used to perform sensitivity analyses to rank the technologies in order of their impact on engine fuel burn and LTO NO_x emissions. Sensitivity values could be + or - in nature. For the current assessment, a positive value indicates that the technology has positive impact on fuel burn (or LTO NO_x) reduction and a negative value has the opposite effect. Technology with the highest absolute sensitivity value is defined to be the most influential technology. The technology with the second highest absolute sensitivity value is the second most influential technology and so on. This approach ranks the technology in the order of their influence on engine performance (i.e., fuel burn or LTO NO_x reductions). The sensitivity information thus obtained from FPI is very useful from the design point of view. For example, engine performance reliability can be improved when uncertainties in the most influential technologies are reduced. Those technologies that do not have significant influences deterministically could nevertheless have strong influences on engine performance reliability if these technologies have huge uncertainties. Weak technology with large uncertainties may have probabilistic sensitivity factors more important than strong technologies with small uncertainties. Unlike deterministic analysis, sensitivity factors in probabilistic analysis are functions of both the deterministic sensitivity and the uncertainty (characterized by the standard deviation).

LTO NOx Definition

The correlation used for the current LTO NO_x calculation is based on flame-tube tests [15,16] and is defined as

$$K(P_{t3})^{0.5945} exp[(T_{t3} - 459.67)(0.002867)] \times (FAR/delphi)^{1.6876}[(1 - P_{t4}/P_{t3})x100]^{-0.5651}$$
 (4)

where

K technology constant

 $\begin{array}{ll} P_{t3} & \text{combustor inlet total pressure} \\ P_{t4} & \text{combustor exit total pressure} \\ T_{t3} & \text{combustor inlet total temperature} \end{array}$

FAR fuel air ratio

delphi 1 – fraction of combustor inlet air used for liner cooling

RESULTS AND DISCUSSION

It is critical to assess the reliability of a new propulsion system because of inherent design uncertainties in the UEET technologies. The current assessment focuses on the technical aspect of engine performance, i.e., mission fuel-burn and LTO NO_x emissions. The results are presented in the form of cumulative distribution functions (CDFs) and probabilistic sensitivities. A CDF gives a relation between a value up to certain magnitude of a response variable (fuel-burn or LTO NOx) and the probability of its occurrence. The results are relative to those of the current state-of-the-art 300-passenger airplane (baseline).

The results show that, the current portfolio of UEET technologies provides a greater than 99% confidence in meeting the UEET Program goal of 70% below the current LTO NO_x rule, while the fuel-burn goal (15% reduction) can be met with 97% confidence. The confidence levels indicate the risk of not meeting the UEET Program goals is very low for the fuel-burn and almost none for the LTO NO_x emissions. The results are shown in Figures 2 and 3.

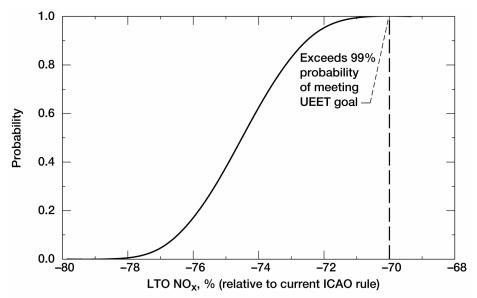


Figure 2.—Cumulative distribution function (CDF) of engine landing and takeoff (LTO) NO_x emissions.

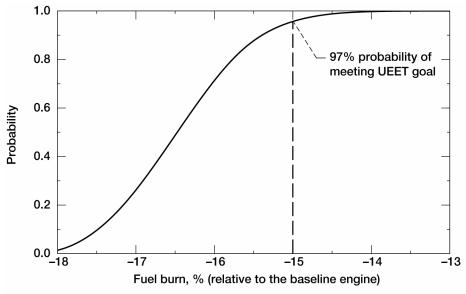


Figure 3.—Cumulative distribution function (CDF) of engine fuel burn.

LTO NO, Sensitivity

The sensitivity of LTO NO_x to the twelve technologies, at 99% probability level is shown in Figure 4. As expected, it shows that the *advanced low-NO_x combustor* (tech-1) has the dominant impact on the LTO NO_x emissions. It implies that to reduce the LTO NO_x emissions beyond the UEET goal, the biggest payoff is to focus on the combustor technology. Other technologies have minimal or no impact on the LTO NO_x emissions. Among all the technologies, the *Advanced compressor disk alloy* (tech-6) is the only one that has a negative impact (though small) on the LTO NO_x . This technology enables the increase in the engine's overall pressure

ratio which also increases the combustor inlet air temperature. That effect alone, assuming all other engine aspects are unchanged, will increase the LTO NO_x emissions.

Fuel Burn Sensitivity

For the fuel-burn reduction, the sensitivity result at 97% is shown in Figure 5. It shows that the *Highly loaded turbine systems* (tech-3), *Highly loaded compressor technology* (tech-2), and the *high Reynolds number design tool for advanced configurations* (tech-9) are the three most influential technologies. The influences of *Advanced compressor disk alloy* (tech-6), *Low-conductivity thermal barrier coating for*

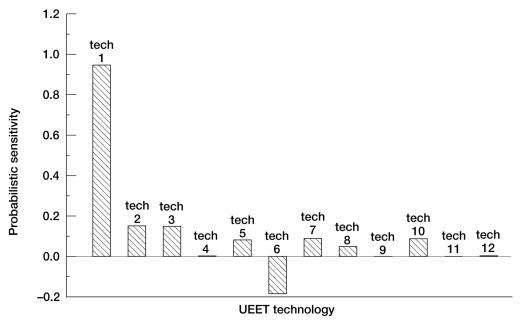


Figure 4.—Sensitivity of engine landing and takeoff (LTO) NO_x emissions.

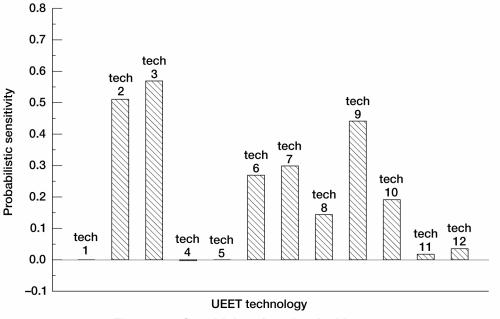


Figure 5.—Sensitivity of engine fuel burn.

turbine airfoils (tech-7), and Rotating machinery clearance management (tech-10) are moderate. Other technologies have minimal or no impact on the fuel-burn reduction.

Among these six top-ranking technologies, tech-6 and tech-7 are material technologies. It is noted that tech-7 and tech-8 (*Advanced airfoil alloy development*) provide the same type of benefit, enable the amount of turbine cooling to be reduced. However, according to the expert opinion (see Table 1), tech-7 enables more cooling flow reduction. As a result, tech-7 has a much higher impact on the fuel burn. Another coolant-reduction technology, tech-4 (*1482 °C CMC turbine vane*), has insignificant impact on the fuel burn, relative to

tech-7 and tech-8. This is because tech-4's coolant reduction comes primarily from the first turbine vane (i.e., non-chargeable cooling) which is not as advantageous as a reduction in chargeable cooling (as for tech-7 and tech-8). Overall, the current results show that advanced materials are the key enablers for meeting the UEET Program goals.

The results also implies that an alternative approach to satisfy the UEET Program goals is to focus on the development of the six largest fuel-burn reduction technologies plus, the $Advanced\ low-NO_x\ combustor$ (tech-1). By reducing the uncertainties of these six technologies by 25%, and by keeping the uncertainty of tech-1 the same, while assuming other

technologies provide only minimum impacts/benefits (from the 3-point estimates), the probability of meeting the UEET fuel-burn goal exceeds 99%, as shown in Figure 6. In fact, this technology portfolio enables a greater than 99% probability of exceeding the UEET fuel-burn goal, with 16.5% fuel-burn reduction relative to the baseline. This is achieved without sacrificing the LTO NO_x goal, as shown in Figure 7.

It is noted that, in a recent independent review of NASA's Aeronautics Technology Programs performed by the National Research Council, these seven technologies, together with tech-8 (*Advanced airfoil alloy development*) have been rated either world-class or exceptionally good technologies [17].

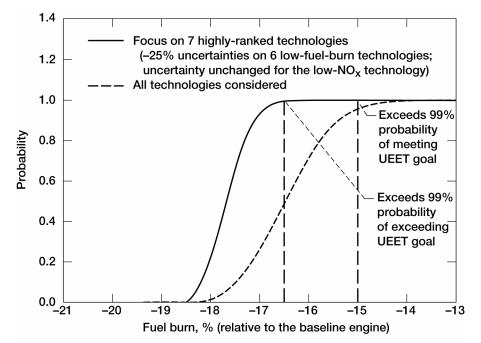


Figure 6.—Impact of uncertainty-reduction on engine fuel burn.

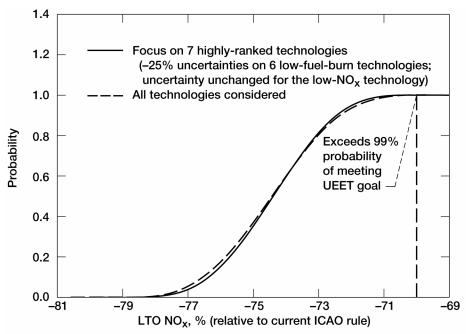


Figure 7.—Impact of uncertainty-reduction on landing and takeoff (LTO) NO_X emissions.

CONCLUSIONS

Based on the current assessment results, the following conclusions are made:

- (1) A large subsonic aircraft equipped with the UEET technologies has a very high probability of meeting the UEET Program goals for fuel burn and LTO NO_x reductions, over 96%.
- (2) The top-seven UEET technologies for fuel-burn or CO₂ reduction are:
 - a. Highly loaded turbine system
 - b. Highly loaded compressor
 - c. High Reynolds number design tool for advanced configuration
 - d. Low conductivity ceramic thermal barrier coating for turbine airfoils
 - e. Advanced compressor disk alloy
 - f. Rotating machinery clearance management
 - g. Advanced airfoil alloy development
- (3) A technology that enables significant non-chargeable coolant reduction (such as 1482 °C CMC turbine vane) is not as advantageous as those that enable significant chargeable coolant reduction (such as Low thermal conductivity ceramic TBC for turbine airfoils and Advanced airfoil alloy development), for fuel burn reduction.
- (4) The Advanced low NO_x combustor technology has the most and dominant impact on the LTO NO_x reductions.
- (5) Advanced materials are key enablers for meeting the UEET Program goals.
- (6) The UEET Program goals can be exceeded with high probability by focusing on the top-ranking technologies and reducing their uncertainties.
- (7) An effective expert opinion elicitation process, or technology audit, is crucial for performing probabilistic assessment. A process that includes both the experts from NASA and the engine industry will ensure the audited data are indeed reasonable representation of each of the technologies' potential.
- (8) The probabilistic approach provides a more realistic and systematic way to assess advanced propulsion technologies, because it accounts for their inherent uncertainties.

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